

Random Access and Source-Channel Coding Error Exponents for Multiple Access Channels

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Abstract

New universal coding/decoding scheme for Random Access with collision detection is given in case of two senders. The result is used to give an achievable source-channel coding error exponent for Multiple-Access channels in case of independent sources.

1 Introduction

This paper addresses a version of the random access model of Luo, Epremidis [6] and Wang, Luo [8], which is similar to the model studied for one-way channels by Csiszár [2]. In the terminology of this paper, in [2] the performance of a codebook library consisting of several constant composition codebooks with pre-determined rates has been analyzed. That result shows that it is possible to achieve universally the same error exponent for each codebook as the random-coding error exponent of this codebook alone. This theorem is used in [2] to give an achievable error exponent for joint source-channel coding (JSCC).

This paper generalized the mentioned results of [2] to (discrete memoryless) multiple-access channels (MACs). A two-senders random-access model is introduced, in which the senders have codebook libraries with constant composition codebooks for multiple rate choices. The error exponent of Liu and Hughes [5] for an individual codebook pair is shown to be universally achievable for each codebook pair in the codebook libraries, supplemented with collision detection in the sense of [6, 8]. In particular, a positive answer is given to the question in [6] whether the results there are still valid if the receiver does not know the channel. Moreover, an achievable JSCC error exponent for transmitting independent sources over a MAC is given, better than that achievable by separate source and channel coding.

Nazari, Anastasopoulos, and Pradhan in [7] derive achievable error exponents for MAC's using α -decoding rule introduced for one-way channels in [3] by Csiszár and Körner. In the present paper a particular α -decoder is used. However, as the proofs follow closely [7], it can be seen that other choices could also be appropriate depending on actual assumptions on the analyzed models.

Note that, an other multiterminal generalization of the JSCC result in [2] appears in Zhong, Alajaji, Campbell [9]. We also mention that, this paper as [2], also has connections with the topic of unequal protection of messages, see for example Borade, Nakiboglu, Zheng [1].

2 Preliminaries, notation

Denote the set $\{1, 2, \dots, M\}$ by $[M]$. The notation follows [4] and [7] whenever possible, for example, the following notations are used: $\mathcal{P}(\mathcal{X} \times \mathcal{Y})$, $\mathcal{P}(\mathcal{X}|\mathcal{U})$, $\mathcal{P}^n(\mathcal{X})$, T_P^n , $T_{P_{X|U}}^n(\mathbf{u})$. Let $\mathcal{P}^n(\mathcal{X}|P_U)$ be the collection of all conditional distribution $V_{X|U}$ for which there exists an $\mathbf{x} \in T_{V_{X|U}}^n(\mathbf{u})$ for some $\mathbf{u} \in T_{P_U}^n$.

Denote $H_V(X, Y)$, $H_V(X, Y, U)$, $I_V(X \wedge Y)$ etc. the entropy and mutual information when the random variables X, Y, U have joint distribution V_{XY} , V_{XYU} etc. Denote $I(\mathbf{x} \wedge \mathbf{y})$, $H(\mathbf{x}, \mathbf{y})$ etc. the information quantities $I_V(X \wedge Y)$, $H_V(X, Y)$ etc. with V_{XY} equal to the type of (\mathbf{x}, \mathbf{y}) . If V_{XYZU} is a multivariate distribution on $\mathcal{X} \times \mathcal{Y} \times \mathcal{Z} \times \mathcal{U}$ then V_{XYU} , V_{XU} , V_{YU} etc. denote the marginal distributions respectively. Moreover, we define the following multi-information (see [4] exercise 3.9. and [5], but different from Yeung's notation):

$$\begin{aligned} I(X_1 \wedge X_2 \wedge \dots \wedge X_N|Y) &\triangleq H(X_1|Y) + H(X_2|Y) + \dots \\ &+ H(X_N|Y) - H(X_1, X_2, \dots, X_N|Y) \end{aligned} \quad (1)$$

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Given a MAC $W : \mathcal{X} \times \mathcal{Y} \rightarrow \mathcal{Z}$, the pentagon

$$\left\{ \begin{array}{l} (R_1, R_2) : 0 \leq R_1 \leq I(X \wedge Z|Y, U), \\ 0 \leq R_2 \leq I(Y \wedge Z|X, U), R_1 + R_2 \leq I(X, Y \wedge Z|U) \end{array} \right\},$$

where U, X, Y, Z have joint distribution equal to $P_U P_{X|U} P_{Y|U} W$, is denoted by $C[W, P_U, P_{X|U}, P_{Y|U}]$. The union of these pentagons, i.e., the capacity region of the MAC W , is denoted by $C(W)$.

3 Random Access with collision detection

In this model two transmitters try to communicate over a MAC W with one common receiver. The channel W is unknown to the senders and may also be unknown to the receiver (but see Remark 2). Both senders have multiple codebooks with block length n . We assume that the codewords' conditional type on an auxiliary sequence \mathbf{u} is fixed in all codebooks, however this common type can vary from codebook to codebook.

Definition 1. Let a finite set \mathcal{U} , a sequence $\mathbf{u} \in \mathcal{U}^n$ of type $P_U \in \mathcal{P}^n(\mathcal{U})$, positive integers M_1 and M_2 , conditional distributions $\{P_{X|U}^i \in \mathcal{P}^n(\mathcal{X}|P_U), i \in [M_1]\}$, $\{P_{Y|U}^j \in \mathcal{P}^n(\mathcal{Y}|P_U), j \in [M_2]\}$, rates $\{R_1^i, i \in [M_1]\}$ and $\{R_2^j, j \in [M_2]\}$ be given parameters. A constant composition codebook library pair of length n with the above parameters is a pair $(\mathcal{A}, \mathcal{B})$ where \mathcal{A} and \mathcal{B} consist of constant composition codebooks (A^1, \dots, A^{M_1}) resp. (B^1, \dots, B^{M_2}) such that $A^i = \{\mathbf{x}_1^i, \mathbf{x}_2^i, \dots, \mathbf{x}_{N_1^i}^i\}$ and $B^j = \{\mathbf{y}_1^j, \mathbf{y}_2^j, \dots, \mathbf{y}_{N_2^j}^j\}$ with $\mathbf{x}_a^i \in \mathcal{T}_{P_{X|U}^i}^n(\mathbf{u})$ and $\mathbf{y}_b^j \in \mathcal{T}_{P_{Y|U}^j}^n(\mathbf{u})$, $i \in [M_1]$, $j \in [M_2]$, $N_1^i = \lfloor e^{nR_1^i} \rfloor$, $N_2^j = \lfloor e^{nR_2^j} \rfloor$, $a \in [N_1^i]$, $b \in [N_2^j]$.

Before sending messages, each transmitter chooses one of its codebooks independently from the other sender. Denote this selection by $(i, j) \in [M_1] \times [M_2]$. The transmitters do not share the result of their selections with each other, neither with the receiver. The senders send codewords $\mathbf{x}_a^i, \mathbf{x}_b^j$. The decoder output $\hat{\mathbf{m}}$ is either a quadruple (i, a, j, b) or "collision". The receiver is required to decode quadruple (i, a, j, b) if the rate pair (R_1^i, R_2^j) of the chosen codebooks is in the interior of $C[W, P_U, P_{X|U}^i, P_{Y|U}^j]$ and to declare "collision" otherwise. Hence, two types of error are defined.

Definition 2. For the codebooks (A^i, B^j) , the average decoding error probability is

$$Err_d(i, j) = \frac{1}{N_1^i N_2^j} \sum_{\mathbf{m} \in A^i \times B^j} Pr\{\hat{\mathbf{m}} \neq \mathbf{m} | \mathbf{m} \text{ is sent}\},$$

and the average collision declaration error probability is

$$Err_c(i, j) = \frac{\sum_{\mathbf{m} \in A^i \times B^j} Pr\{\hat{\mathbf{m}} \neq \text{"collision"} | \mathbf{m} \text{ is sent}\}}{N_1^i N_2^j}.$$

To state our basic theorem we need the following notions from [5]; the index HL refers to the authors of [5].

$$\begin{aligned} \mathcal{V}_{HL} &= \mathcal{V}_{HL}(W, P_U, P_{X|U}, P_{Y|U}) \\ &\triangleq \{V_{UXYZ} : V_{UX} = P_U P_{X|U}, V_{UY} = P_U P_{Y|U}\} \end{aligned} \quad (2)$$

$$\begin{aligned} EX_{HL}(R_1, R_2, W, P_U, P_{X|U}, P_{Y|U}) &\triangleq \min_{V_{UXYZ} \in \mathcal{V}_{HL}} [D(V_{Z|XYU} || W | V_{XYU}) \\ &\quad + I_V(X \wedge Y|U) + |I_V(X \wedge YZ|U) - R_1|^+] \end{aligned} \quad (3)$$

$$\begin{aligned} EY_{HL}(R_1, R_2, W, P_U, P_{X|U}, P_{Y|U}) &\triangleq \min_{V_{UXYZ} \in \mathcal{V}_{HL}} [D(V_{Z|XYU} || W | V_{XYU}) \\ &\quad + I_V(X \wedge Y|U) + |I_V(Y \wedge XZ|U) - R_2|^+] \end{aligned} \quad (4)$$

$$\begin{aligned} EXY_{HL}(R_1, R_2, W, P_U, P_{X|U}, P_{Y|U}) &\triangleq \min_{V_{UXYZ} \in \mathcal{V}_{HL}} [D(V_{Z|XYU} || W | V_{XYU}) \\ &\quad + I_V(X \wedge Y|U) + |I_V(X \wedge Y \wedge Z|U) - R_1 - R_2|^+] \end{aligned} \quad (5)$$

$$\begin{aligned} E_{HL}(R_1, R_2, W, P_U, P_{X|U}, P_{Y|U}) &\triangleq \min\{EX_{HL}, EY_{HL}, EXY_{HL}\} \end{aligned} \quad (6)$$

Theorem 1 shows that the error exponent of [5] for an individual codebook pair is achievable for this general setting.

Theorem 1. For each n let constant composition random access codebook library parameters as in Definition 1 be given with a common set \mathcal{U} and with $\frac{1}{n}\log M_1 \rightarrow 0$, $\frac{1}{n}\log M_2 \rightarrow 0$ as $n \rightarrow \infty$. Then there exist a sequence $\delta_n(|\mathcal{U}|, |\mathcal{X}|, |\mathcal{Y}|, |\mathcal{Z}|, \{M_1\}_{n=1}^\infty, \{M_2\}_{n=1}^\infty) \rightarrow 0$ and for each n a constant composition codebook-library pair $(\mathcal{A}, \mathcal{B})$ with the given parameters, and decoder mappings with the following properties:

(i) For all $(i, j) \in [M_1] \times [M_2]$

$$\text{Err}_d(i, j) \leq e^{-n(E_{HL}(R_1^i, R_2^j, W, P_U, P_{X|U}^i, P_{Y|U}^j) - \delta_n)}. \quad (7)$$

(ii) If (R_1^i, R_2^j) is not in the interior of $C[W, P_U, P_{X|U}^i, P_{Y|U}^j]$ then

$$\text{Err}_c(i, j) < \delta_n. \quad (8)$$

Remark 1. Note that in part (i) of Theorem 1 $E_{HL}(R_1^i, R_2^j, W, P_U, P_{X|U}^i, P_{Y|U}^j) > 0$ iff (R_1^i, R_2^j) is in the interior of $C[W, P_U, P_{X|U}^i, P_{Y|U}^j]$.

The next packing lemma is an extension of Lemma 4 in [7] for this multiple codebooks setting, it provides the appropriate codebook library pair for Theorem 1. Note that multi information is used instead of the original notation of [7].

Lemma 1. Let a sequence of constant composition random access codebook library parameters be given as in Theorem 1. Then there exist a sequence $\delta'_n(|\mathcal{U}|, |\mathcal{X}|, |\mathcal{Y}|, \{M_1\}_{n=1}^\infty, \{M_2\}_{n=1}^\infty) \rightarrow 0$ and for each n a constant composition codebook-library pair $(\mathcal{A}, \mathcal{B})$ with the given parameters such that for any $(i, k) \in [M_1]^2$ and $(j, l) \in [M_2]^2$ and for all $V_{UX\hat{X}Y\hat{Y}U} \in \mathcal{P}^n(\mathcal{U} \times \mathcal{X} \times \mathcal{X} \times \mathcal{Y} \times \mathcal{Y})$:

$$\begin{aligned} K_{k,l}^{i,j}[V_{UXY}] &\leq 2^{-n(I_V(X \wedge Y|U) - R_1^i - R_2^j - \delta'_n)} \\ K_{k,l}^{i,j}[V_{UX\hat{X}Y}] &\leq 2^{-n(I_V(X \wedge \hat{X} \wedge Y|U) - R_1^i - R_2^j - R_1^k - \delta'_n)} \\ K_{k,l}^{i,j}[V_{UXY\hat{Y}}] &\leq 2^{-n(I_V(X \wedge Y \wedge \hat{Y}|U) - R_1^i - R_2^j - R_2^l - \delta'_n)} \\ K_{k,l}^{i,j}[V_{UX\hat{X}Y\hat{Y}}] &\leq 2^{-n(I_V(X \wedge \hat{X} \wedge Y \wedge \hat{Y}|U) - R_1^i - R_2^j - R_1^k - R_2^l - \delta'_n)}, \end{aligned}$$

where

$$K_{k,l}^{i,j}[V_{UXY}] \triangleq \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} \mathbb{1}_{\mathcal{T}_{V_{UXY}}}(\mathbf{u}, \mathbf{x}_a^i, \mathbf{y}_b^j) \quad (9)$$

$$K_{k,l}^{i,j}[V_{UX\hat{X}Y}] \triangleq \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} \sum_{\substack{c=1 \\ c \neq a \text{ if } i=k}}^{N_1^k} \mathbb{1}_{\mathcal{T}_{V_{UX\hat{X}Y}}}(\mathbf{u}, \mathbf{x}_a^i, \mathbf{x}_c^k, \mathbf{y}_b^j) \quad (10)$$

$$K_{k,l}^{i,j}[V_{UXY\hat{Y}}] \triangleq \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} \sum_{\substack{d=1 \\ d \neq b \text{ if } j=l}}^{N_2^l} \mathbb{1}_{\mathcal{T}_{V_{UXY\hat{Y}}}}(\mathbf{u}, \mathbf{x}_a^i, \mathbf{y}_b^j, \mathbf{y}_d^l) \quad (11)$$

$$\begin{aligned} K_{k,l}^{i,j}[V_{UX\hat{X}Y\hat{Y}}] &\triangleq \\ &\triangleq \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} \sum_{\substack{c=1 \\ c \neq a \text{ if } i=k}}^{N_1^k} \sum_{\substack{d=1 \\ d \neq b \text{ if } j=l}}^{N_2^l} \mathbb{1}_{\mathcal{T}_{V_{UX\hat{X}Y\hat{Y}}}}(\mathbf{u}, \mathbf{x}_a^i, \mathbf{x}_c^k, \mathbf{y}_b^j, \mathbf{y}_d^l). \end{aligned} \quad (12)$$

Sketch of proof: Let $(\mathcal{A}, \mathcal{B})$ be a random constant composition codebook library pair, i. e for all $i \in [M_1], j \in [M_2]$ the codewords of A^i, B^j are chosen independently and uniformly from $\mathcal{T}_{P_{X|U}^i}^n(\mathbf{u})$ and $\mathcal{T}_{P_{Y|U}^j}^n(\mathbf{u})$ respectively. Let T be the following random variable:

$$\begin{aligned} T &= \sum_{i=1}^{M_1} \sum_{j=1}^{M_2} \sum_{k=1}^{M_1} \sum_{l=1}^{M_2} \sum_{\substack{V_{UX\hat{X}Y\hat{Y}U} \in \\ \mathcal{P}^n(\mathcal{U} \times \mathcal{X} \times \mathcal{X} \times \mathcal{Y} \times \mathcal{Y})}} \left[K_{k,l}^{i,j}[V_{UXY}] \right. \\ &\quad \cdot 2^{n(I_V(X \wedge Y|U) - R_1^i - R_2^j - \delta'_n)} \\ &\quad + K_{k,l}^{i,j}[V_{UX\hat{X}Y}] 2^{n(I_V(X \wedge \hat{X} \wedge Y|U) - R_1^i - R_2^j - R_1^k - \delta'_n)} \\ &\quad + K_{k,l}^{i,j}[V_{UXY\hat{Y}}] 2^{n(I_V(X \wedge Y \wedge \hat{Y}|U) - R_1^i - R_2^j - R_2^l - \delta'_n)} \\ &\quad + K_{k,l}^{i,j}[V_{UX\hat{X}Y\hat{Y}}] 2^{n(I_V(X \wedge \hat{X} \wedge Y \wedge \hat{Y}|U) - R_1^i - R_2^j - R_1^k - R_2^l - \delta'_n)} \\ &\quad \left. \cdot 2^{n(-R_2^l - \delta'_n)} \right]. \end{aligned} \quad (13)$$

Using basic properties of types (similarly as in [2, 7]) it can be seen that we can choose $\delta'_n \rightarrow 0$ such that $\mathbb{E}[T] < 1$. Hence there exists a realization of the codebook library pair with $T < 1$. Taking into account the positivity of the terms of T , the lemma is proved. ■

Sketch of proof of Theorem 1: Lemma 1 provides the appropriate constant composition codebook-library pair $(\mathcal{A}, \mathcal{B})$. To construct the decoder, define $\alpha : \mathcal{P}(\mathcal{U} \times \mathcal{X} \times \mathcal{Y} \times \mathcal{Z}) \rightarrow R$ by $\alpha(V_{UXYZ}) = I_V(X \wedge Y \wedge Z|U)$. In the first stage of decoding, the receiver determine the indices $\hat{k} \in [M_1]$, $\hat{l} \in [M_2]$, $\hat{a} \in [N_1^{\hat{k}}]$, $\hat{b} \in [N_2^{\hat{l}}]$ which maximize $\alpha(\mathbf{u}, \mathbf{x}_a^{\hat{k}}, \mathbf{y}_b^{\hat{l}}, \mathbf{z}) - R_1^{\hat{k}} - R_2^{\hat{l}}$, where \mathbf{z} denotes the output sequence and α is evaluated on the joint type of $(\mathbf{u}, \mathbf{x}_a^{\hat{k}}, \mathbf{y}_b^{\hat{l}}, \mathbf{z})$. In the second stage, to deal with collisions, the decoder checks the following three inequalities:

$$I(\mathbf{x}_a^{\hat{k}} \wedge \mathbf{y}_b^{\hat{l}} \wedge \mathbf{z}|\mathbf{u}) - R_1^{\hat{k}} - R_2^{\hat{l}} > \eta_n \quad (14)$$

$$I(\mathbf{x}_a^{\hat{k}} \wedge \mathbf{y}_b^{\hat{l}}, \mathbf{z}|\mathbf{u}) - R_1^{\hat{k}} > \eta_n \quad (15)$$

$$I(\mathbf{y}_b^{\hat{l}} \wedge \mathbf{x}_a^{\hat{k}}, \mathbf{z}|\mathbf{u}) - R_2^{\hat{l}} > \eta_n, \quad (16)$$

where $\eta_n(|\mathcal{U}|, |\mathcal{X}|, |\mathcal{Y}|, |\mathcal{Z}|, \{M_1\}_{n=1}^\infty, \{M_2\}_{n=1}^\infty) \rightarrow 0$ is an appropriately chosen positive sequence. If the above three inequalities are fulfilled then the decoder decodes $\mathbf{x}_a^{\hat{k}}, \mathbf{y}_b^{\hat{l}}$ as the codewords sent, if at least one of them is not fulfilled, then the decoder reports “collision”.

Some details about the necessary calculations can be found in the Appendix. ■

Remark 2. Other α -decoders can be also used; if the receiver knows the channel W , the α function can depend on W . Note also that for the sake of brevity the expurgation method for multiple-access channel in [7] is not used in this paper. However, it is possible to prove an expurgated version of Lemma 1 which leads larger achievable error exponent for small rates.

4 Source-channel coding

Let two independent discrete memoryless sources (DMS) Q_1, Q_2 with alphabets $\mathcal{S}_1, \mathcal{S}_2$ be given. We want to transmit these sources over MAC W . It is assumed that the sources Q_1 and Q_2 and the channel W are known by the encoders, however, not known by the decoder.

Definition 3. A source-channel code of length n is a mapping triple (f_1, f_2, φ) with encoders $f_1 : \mathcal{S}_1^n \rightarrow \mathcal{X}^n$, $f_2 : \mathcal{S}_2^n \rightarrow \mathcal{Y}^n$ and decoder $\varphi : \mathcal{Z}^n \rightarrow \mathcal{S}_1^n \times \mathcal{S}_2^n$.

Definition 4. The error of a source-channel code (f_1, f_2, φ) of length n is defined by

$$Err(f_1, f_2, \varphi) = \sum_{\substack{(\mathbf{s}_1, \mathbf{s}_2) \in \\ \mathcal{S}_1^n \times \mathcal{S}_2^n}} Q_1^n(\mathbf{s}_1) Q_2^n(\mathbf{s}_2) p_e(\mathbf{s}_1, \mathbf{s}_2), \quad (17)$$

where

$$p_e(\mathbf{s}_1, \mathbf{s}_2) = W^n(\{\mathbf{z} \in \mathcal{Z}^n : \varphi(\mathbf{z}) \neq (\mathbf{s}_1, \mathbf{s}_2)\} | f_1(\mathbf{s}_1), f_2(\mathbf{s}_2)).$$

We assume that $(H(Q_1), H(Q_2))$ is in the interior of $C(W)$. In this case Q_1, Q_2 can be reliably transmitted over channel W , for example, by separate source and channel coding. Regarding error exponents, by separate coding the exponent $Es_{HL}(Q_1, Q_2, W)$ is achievable, where

$$Es_{HL}(Q_1, Q_2, W) \triangleq \max_{\substack{0 \leq R_1 \leq \log|\mathcal{S}_1| \\ 0 \leq R_2 \leq \log|\mathcal{S}_2|}} \min \left[e_1(R_1, Q_1), e_2(R_2, Q_2), E_{HL}(R_1, R_2, W) \right], \quad (18)$$

$e_1(R_1, Q_1), e_2(R_2, Q_2)$ are the source reliability functions

$$e_i(R_i, Q_i) \triangleq \min_{P: H(P) \geq R_i} D(P||Q_i), \quad i \in \{1, 2\}, \quad (19)$$

and

$$E_{HL}(R_1, R_2, W) \triangleq \sup_{\substack{\mathcal{U} \\ P_U \in \mathcal{P}(\mathcal{U})}} \sup_{\substack{P_{X|U} \in \mathcal{P}(\mathcal{X}|\mathcal{U}) \\ P_{Y|U} \in \mathcal{P}(\mathcal{Y}|\mathcal{U})}} E_{HL}[R_1, R_2, W, P_U, P_{X|U}, P_{Y|U}]. \quad (20)$$

Note that $Es_{HL}(Q_1, Q_2, W) > 0$. Of course, (18) could be improved replacing $E_{HL}(R_1, R_2, W)$ by the reliability function of channel W which, however, is not known in general.

In this section a larger exponent than E_{HL} is given using JSCC.

For arbitrary \mathcal{U} and $P_U \in \mathcal{P}(\mathcal{U})$ let $G_1(\mathcal{U})$ and $G_2(\mathcal{U})$ be the set of all continuous mappings $[0, \log|S_1|] \rightarrow \mathcal{P}(\mathcal{X}|\mathcal{U})$ and $[0, \log|S_2|] \rightarrow \mathcal{P}(\mathcal{Y}|\mathcal{U})$ respectively, and define

$$Ej(Q_1, Q_2, W) \triangleq \sup_{\substack{\mathcal{U} \\ P_U \in \mathcal{P}(\mathcal{U})}} \sup_{\substack{g_1 \in G_1(\mathcal{U}) \\ g_2 \in G_2(\mathcal{U})}} Ej(Q_1, Q_2, W, P_U, g_1, g_2)$$

where

$$\begin{aligned} Ej(Q_1, Q_2, W, P_U, g_1, g_2) \triangleq & \min_{\substack{0 \leq R_1 \leq \log|S_1| \\ 0 \leq R_2 \leq \log|S_2|}} [e_1(R_1) + e_2(R_2) \\ & + E_{HL}(R_1, R_2, W, P_U, g_1(R_1), g_2(R_2))]. \end{aligned} \quad (21)$$

Proposition 2. $E_{HL}(Q_1, Q_2, W) \leq Ej(Q_1, Q_2, W)$.

Proof: Restricting the supremum to constant functions g_1, g_2 in (21), we see that:

$$\begin{aligned} Ej(Q_1, Q_2, W) \geq & \sup_{\substack{\mathcal{U} \\ P_U \in \mathcal{P}(\mathcal{U})}} \sup_{\substack{P_{X|U} \in \mathcal{P}(\mathcal{X}|\mathcal{U}) \\ P_{Y|U} \in \mathcal{P}(\mathcal{Y}|\mathcal{U})}} \min_{\substack{0 \leq R_1 \leq \log|S_1| \\ 0 \leq R_2 \leq \log|S_2|}} [e_1(R_1) \\ & + e_2(R_2) + E_{HL}(R_1, R_2, W, P_U, P_{X|U}, P_{Y|U})]. \end{aligned} \quad (22)$$

Using the definition of $e_1(R_1)$, $e_2(R_2)$ and $E_{HL}(R_1, R_2, W, P_U, P_{X|U}, P_{Y|U})$ it can be easily seen that the right-hand side of (22) is greater than or equal to $E_{HL}(Q_1, Q_2, W)$. ■

Remark 3. Actually, the strict inequality holds, except in very special cases.

The following theorem shows that $Ej(Q_1, Q_2, W)$ is an achievable error exponent for this source-channel coding scenario, hence JSCC leads to larger exponent than separate source and channel coding. More exactly, we show that for any choice of P_U, g_1, g_2 , the exponent $Ej(Q_1, Q_2, W, P_U, g_1, g_2)$ is achievable even if the senders do not know the sources and the channel; if they do know them, they can optimize in P_U, g_1, g_2 , to achieve $Ej(Q_1, Q_2, W)$.

Theorem 3. Let $\mathcal{U}, P_U \in \mathcal{P}(\mathcal{U})$, $g_1 \in G_1(\mathcal{U})$ and $g_2 \in G_2(\mathcal{U})$ be given. There exist a source-channel code for each n and a sequence $\nu_n(|S_1|, |S_2|, |\mathcal{U}|, |\mathcal{X}|, |\mathcal{Y}|, |\mathcal{Z}|) \rightarrow 0$ with

$$Err(f_1, f_2, \varphi) \leq 2^{-n(Ej(Q_1, Q_2, W, P_U, g_1, g_2) - \nu_n)}. \quad (23)$$

Sketch of proof: Approximate uniformly P_U, g_1, g_2 by sequences $P_U[n] \in \mathcal{P}(\mathcal{U})$, $g_1[n] : [0, \log|S_1|] \rightarrow \mathcal{P}(\mathcal{X}|P_U(n))$, $g_2[n] : [0, \log|S_2|] \rightarrow \mathcal{P}(\mathcal{Y}|P_U(n))$.

Let $\mathbf{u} \in \mathcal{T}_{P_U[n]}^n$ be arbitrary sequence. Choose $M_1 = |\mathcal{P}^n(S_1)|$ and $M_2 = |\mathcal{P}^n(S_2)|$. Let $P_1^1, P_1^2, \dots, P_1^{M_1}$ and $P_2^1, P_2^2, \dots, P_2^{M_2}$ denote all possible types from $\mathcal{P}^n(S_1)$ and $\mathcal{P}^n(S_2)$ respectively. For all $i \in [M_1]$, $j \in [M_2]$ let N_1^i and N_2^j be equal to $|\mathcal{T}_{P_1^i}^n|$ and $|\mathcal{T}_{P_2^j}^n|$ respectively, and let $P_{X|U}^i$ and $P_{Y|U}^j$ be equal to $g_1[n](N_1^i)$, $g_2[n](N_2^j)$ respectively. Applying Theorem 1 with these parameters consider the resulting codebook library pair $(\mathcal{A}, \mathcal{B})$ and the decoder mapping ϕ satisfying (7) for all $(i, j) \in [M_1] \times [M_2]$.

Let $f_1 : \mathcal{S}_1^n \rightarrow \mathcal{X}^n$ and $f_2 : \mathcal{S}_2^n \rightarrow \mathcal{Y}^n$ be the mappings which map each $\mathcal{T}_{P_1^i}^n$ and $\mathcal{T}_{P_2^j}^n$ to A^i and B^j respectively. Let $\varphi : \mathcal{Z}^n \rightarrow \mathcal{S}_1^n \times \mathcal{S}_2^n$ be the mapping which first determines a codeword pair from $(\mathcal{A}, \mathcal{B})$ using ϕ , then uses the inverse of f_1 and f_2 to determine the source sequences. The crucial step is the following equation

$$\begin{aligned} Err(f_1, f_2, \varphi) = & \sum_{i=1}^{M_1} \sum_{j=1}^{M_2} Q_1^n(\mathcal{T}_{P_1^i}^n) Q_2^n(\mathcal{T}_{P_2^j}^n) \\ & \cdot \frac{1}{|\mathcal{T}_{P_1^i}^n|} \frac{1}{|\mathcal{T}_{P_2^j}^n|} \sum_{\mathbf{s}_1 \in \mathcal{T}_{P_1^i}^n} \sum_{\mathbf{s}_2 \in \mathcal{T}_{P_2^j}^n} p_e(\mathbf{s}_1, \mathbf{s}_2) \end{aligned} \quad (24)$$

Note that the second line of (24) is $Err_d(i, j)$ in the terminology of Theorem 1. Hence substituting (7) into (24) and using (19) and standard properties of types, this theorem is proved. ■

Remark 4. Analogously to Lemma 2 of [2], it can be shown that the error exponent can not be greater than

$$\min_{\substack{0 \leq R_1 \leq \log|S_1| \\ 0 \leq R_2 \leq \log|S_2|}} [e_1(R_1) + e_2(R_2) + E(R_1, R_2, W)]$$

where $E(R_1, R_2, W)$ is the reliability function of channel W .

[Sketch of proof of Theorem 1] Let us define the following sets for all $i \in [M_1]$, $j \in [M_2]$, $a \in [N_1^i]$, $b \in [N_2^j]$:

$$D_{a,b}^{i,j} \triangleq \left\{ \mathbf{z} : \begin{aligned} &\alpha(\mathbf{u}, \mathbf{x}_a^i, \mathbf{y}_b^j, \mathbf{z}) - R_1^i - R_2^j \\ &\geq \alpha(\mathbf{u}, \mathbf{x}_c^k, \mathbf{y}_d^l, \mathbf{z}) - R_1^k - R_2^l, \text{ for all} \\ &k \in [M_1], l \in [M_2], c \in [N_1^k], d \in [N_2^l] \end{aligned} \right\} \quad (25)$$

$$O_{a,b}^{i,j} \triangleq \left\{ \mathbf{z} : \begin{aligned} &\mathbf{I}(\mathbf{x}_a^i \wedge \mathbf{y}_b^j \wedge \mathbf{z} | \mathbf{u}) - R_1^i - R_2^j \leq \eta_n \text{ or} \\ &\mathbf{I}(\mathbf{x}_a^i \wedge \mathbf{y}_b^j, \mathbf{z} | \mathbf{u}) - R_1^i \leq \eta_n \text{ or} \\ &\mathbf{I}(\mathbf{y}_b^j \wedge \mathbf{x}_a^i, \mathbf{z} | \mathbf{u}) - R_2^j \leq \eta_n \end{aligned} \right\} \quad (26)$$

Then for all $(i, j) \in [M_1] \times [M_2]$:

$$\begin{aligned} Err_d(i, j) &\leq \frac{1}{N_1^i N_2^j} \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} W^n \left(O_{a,b}^{i,j} | \mathbf{x}_a^i, \mathbf{y}_b^j \right) \\ &+ \frac{1}{N_1^i N_2^j} \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} W^n \left(\bigcup_{\substack{c=1 \\ c \neq a}}^{N_1^i} D_{c,b}^{i,j} | \mathbf{x}_a^i, \mathbf{y}_b^j \right) \\ &+ \frac{1}{N_1^i N_2^j} \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} W^n \left(\bigcup_{\substack{d=1 \\ d \neq b}}^{N_2^j} D_{a,d}^{i,j} | \mathbf{x}_a^i, \mathbf{y}_b^j \right) \\ &+ \frac{1}{N_1^i N_2^j} \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} W^n \left(\bigcup_{\substack{c=1 \\ c \neq a}}^{N_1^i} \bigcup_{\substack{d=1 \\ d \neq b}}^{N_2^j} D_{c,d}^{i,j} | \mathbf{x}_a^i, \mathbf{y}_b^j \right) \\ &+ \sum_{\substack{k=1 \\ k \neq i}}^{M_1} \frac{1}{N_1^i N_2^j} \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} W^n \left(\bigcup_{c=1}^{N_1^k} D_{c,b}^{k,j} | \mathbf{x}_a^i, \mathbf{y}_b^j \right) \\ &+ \sum_{\substack{l=1 \\ l \neq j}}^{M_2} \frac{1}{N_1^i N_2^j} \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} W^n \left(\bigcup_{d=1}^{N_2^l} D_{a,d}^{i,l} | \mathbf{x}_a^i, \mathbf{y}_b^j \right) \\ &+ \sum_{\substack{k=1 \\ k \neq i}}^{M_1} \sum_{\substack{l=1 \\ l \neq j}}^{M_2} \frac{1}{N_1^i N_2^j} \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} W^n \left(\bigcup_{c=1}^{N_1^k} \bigcup_{d=1}^{N_2^l} D_{c,d}^{k,l} | \mathbf{x}_a^i, \mathbf{y}_b^j \right). \end{aligned} \quad (27)$$

For the sake of brevity, we introduce the following notations for the terms of the right-hand side of equation (27):

$$\begin{aligned} Err_d(i, j) &\leq th^{i,j} + errorX_{i,j}^{i,j} + errorY_{i,j}^{i,j} \\ &+ errorXY_{i,j}^{i,j} + \sum_{\substack{k=1 \\ k \neq i}}^{M_1} errorX_{k,j}^{i,j} \\ &+ \sum_{\substack{l=1 \\ l \neq j}}^{M_2} errorY_{i,l}^{i,j} + \sum_{\substack{k=1 \\ k \neq i}}^{M_1} \sum_{\substack{l=1 \\ l \neq j}}^{M_2} errorXY_{k,l}^{i,j}. \end{aligned} \quad (28)$$

Let us define the following expressions for all $i \in [M_1]$, $j \in [M_2]$, $k \in [M_1]$, $l \in [M_2]$:

$$\mathcal{V}_{k,l}^{i,j} \triangleq \left\{ \begin{aligned} &V_{UXY\tilde{X}Z} : \\ &\alpha(V_{UXYZ}) - R_1^i \leq \alpha(V_{U\tilde{X}YZ}) - R_1^k, \\ &V_{UX} = P_U P_{X|U}^i, \quad V_{U\tilde{X}} = P_U P_{X|U}^k, \\ &V_{UY} = P_U P_{Y|U}^j \end{aligned} \right\} \quad (29)$$

$$\mathcal{V}_{k,l}^{i,j} \triangleq \left\{ \begin{aligned} &V_{UXY\tilde{Y}Z} : \\ &\alpha(V_{UXYZ}) - R_2^j \leq \alpha(V_{UX\tilde{Y}Z}) - R_2^l, \\ &V_{UX} = P_U P_{X|U}^i, \quad V_{UY} = P_U P_{Y|U}^j, \\ &V_{U\tilde{Y}} = P_U P_{Y|U}^l. \end{aligned} \right\} \quad (30)$$

$$\mathcal{V}\mathcal{X}\mathcal{Y}_{k,l}^{i,j} \triangleq \left\{ \begin{array}{l} V_{UXY\tilde{X}\tilde{Y}Z} : \\ \alpha(V_{UXYZ}) - R_1^i - R_2^j \\ \leq \alpha(V_{U\tilde{X}\tilde{Y}Z}) - R_1^k - R_2^l, \\ V_{UX} = P_U P_{X|U}^i, \quad V_{U\tilde{X}} = P_U P_{X|U}^k, \\ V_{UY} = P_U P_{Y|U}^j, \quad V_{U\tilde{Y}} = P_U P_{Y|U}^l. \end{array} \right\} \quad (31)$$

$$EX_{k,l}^{i,j} \triangleq \min_{V_{UXY\tilde{X}\tilde{Y}Z} \in \mathcal{V}\mathcal{X}_{k,l}^{i,j}} D(V_{Z|XYU} || W | V_{XYU}) + I_V(X \wedge Y | U) + |I_V(\tilde{X} \wedge XYZ | U) - R_1^i|^+ \quad (32)$$

$$EY_{k,l}^{i,j} \triangleq \min_{V_{UXY\tilde{Y}Z} \in \mathcal{V}\mathcal{Y}_{k,l}^{i,j}} D(V_{Z|XYU} || W | V_{XYU}) + I_V(X \wedge Y | U) + |I_V(\tilde{Y} \wedge XYZ | U) - R_2^j|^+ \quad (33)$$

$$EXY_{k,l}^{i,j} \triangleq \min_{V_{UXY\tilde{X}\tilde{Y}Z} \in \mathcal{V}\mathcal{X}\mathcal{Y}_{k,l}^{i,j}} D(V_{Z|XYU} || W | V_{XYU}) + I_V(X \wedge Y | U) + |I_V(\tilde{X}\tilde{Y} \wedge XYZ | U) + I_V(\tilde{X} \wedge \tilde{Y} | U) - R_1^i - R_2^j|^+ \quad (34)$$

Relating the error probabilities to packing functions (9)-(12) as in [7] gives:

$$\begin{aligned} \text{error}X_{k,l}^{i,j} &\leq 2^{-n(EX_{k,l}^{i,j} - \delta_n'')}, \quad \text{error}Y_{k,l}^{i,j} \leq 2^{-n(EY_{k,l}^{i,j} - \delta_n'')} \\ \text{error}XY_{k,l}^{i,j} &\leq 2^{-n(EXY_{k,l}^{i,j} - \delta_n'')} \end{aligned}$$

for some sequence $\delta_n''(|\mathcal{U}|, |\mathcal{X}|, |\mathcal{Y}|, |\mathcal{Z}|, M_1, M_2) \rightarrow 0$. Moreover, using definitions (2)-(6) and (34) we get:

$$\begin{aligned} EX_{k,l}^{i,j} &\geq EX_{HL}^{i,j}, \quad EY_{k,l}^{i,j} \geq EY_{HL}^{i,j} \\ EXY_{k,l}^{i,j} &\geq EXY_{HL}^{i,j}. \end{aligned}$$

Furthermore, using standard properties of types it follows that $th^{(i,j)} < 2^{-n(ETH(i,j) - \delta_n'')}$ where $ETH(i, j)$ is defined by

$$\min_{V_{UXYZ} \in \mathcal{O}^{i,j}} D(V_{Z|XYU} || W | V_{XYU}) + I_V(X \wedge Y | U) \quad (35)$$

where

$$\mathcal{O}^{i,j} \triangleq \left\{ \begin{array}{l} V_{UXYZ} : V_{UX} = P_U P_{X|U}^i, V_{UY} = P_U P_{Y|U}^j \\ I_V(X \wedge Y, Z | U) - R_1^i \leq \eta_n \text{ or} \\ I_V(Y \wedge X, Z | U) - R_2^j \leq \eta_n \text{ or} \\ I_V(X \wedge Y \wedge Z | U) - R_1^i - R_2^j \leq \eta_n \end{array} \right\}$$

Using that $\min(EX_{HL}^{i,j}, EY_{HL}^{i,j}, EXY_{HL}^{i,j}) \leq ETH(i, j) + \eta_n$, the above inequalities prove part (i) of Theorem 1.

To prove part (ii) the following bound is useful:

$$\begin{aligned} Err_c(i, j) &\leq \frac{1}{N_1^i N_2^j} \sum_{a=1}^{N_1^i} \sum_{b=1}^{N_2^j} W^n(\mathbf{z} : \exists k \in [M_1], \exists l \in [M_2], \\ &\quad \exists c \in [N_1^k], \exists d \in [N_2^l] \text{ such that } \mathbf{z} \notin O_{c,d}^{k,l} | \mathbf{x}_a^i, \mathbf{y}_b^j). \end{aligned} \quad (36)$$

Using union bound, it is possible to expand (36) same way as $Err_c(i, j)$ is expanded in (27). Then the term corresponding to case $(k, c, l, d) = (i, a, j, b)$ can be upper bounded using standard properties of types. This upper bound leads to exponent $\min_{V_{UXYZ} \notin \mathcal{O}^{i,j}} D(V_{Z|XYU} || W | V_{XYU}) + I_V(X \wedge Y | U)$. The other cases can be upper bounded using the technique of [7] by exponent similar to (32)-(34), the sets on which the minimum is taken are different. Using the properties of these sets all terms within the positive part sign $|\dots|^+$ can be lower bounded by η_n . Altogether, it can be seen that $Err_c(i, j)$ is small, if η_n goes to 0 appropriately slow.

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